#### Three.IV Matrix Operations

Linear Algebra
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## Sums and Scalar Products

### Representing operations on linear functions

Recall the operations on linear functions of scalar multiplication and addition defined by: for f, g:  $V \to W$ , where  $r \in \mathbb{R}$  the scalar multiple  $r \cdot f$  function does this

$$\vec{v} \stackrel{r \cdot f}{\longmapsto} r \cdot (f(\vec{v}))$$

and the sum of the two functions does this.

$$\vec{v} \stackrel{f+g}{\longmapsto} f(\vec{v}) + g(\vec{v})$$

We will see how the matrix representations  $\operatorname{Rep}_{B,D}(f)$  and  $\operatorname{Rep}_{B,D}(g)$  combine to give  $\operatorname{Rep}_{B,D}(r \cdot f)$  and  $\operatorname{Rep}_{B,D}(f+g)$ .

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*Example* Suppose that  $f: V \to W$  is the linear map represented with respect to some bases B, D by this matrix.

$$F = \operatorname{Rep}_{B,D}(f) = \begin{pmatrix} 2 & 1 \\ 3 & 4 \end{pmatrix}$$

We will find the matrix representing the function 6f. That is, we will show how to produce the matrix representation of 6f from the matrix representation F of f.

Consider representations of vectors from the domain and codomain.

$$\operatorname{Rep}_{\mathrm{B}}(\vec{\mathsf{v}}) = \begin{pmatrix} \mathsf{v}_1 \\ \mathsf{v}_2 \end{pmatrix} \quad \operatorname{Rep}_{\mathrm{D}}(\vec{\mathsf{w}}) = \begin{pmatrix} \mathsf{w}_1 \\ \mathsf{w}_2 \end{pmatrix}$$

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Since  $6\vec{w} = 6 \cdot (w_1\vec{\delta}_1 + w_2\vec{\delta}_2) = 6w_1\vec{\delta}_1 + 6w_2\vec{\delta}_2$ , application of the function 6f:  $V \to W$  yields this representation of the codomain vector.

$$\operatorname{Rep}_{\mathbf{D}}(6f(\vec{\mathbf{v}})) = \operatorname{Rep}_{\mathbf{D}}(6\vec{\mathbf{w}}) = \begin{pmatrix} 6w_1 \\ 6w_2 \end{pmatrix}$$

So moving from the function f to the function 6f has the effect on the representations of vectors that it transforms the codomain vector by multiplying each entry by 6.

In this example the action of the function is represented as here.

$$\operatorname{Rep}_{B,D}(f) \cdot \operatorname{Rep}_{B}(\vec{v}) = \begin{pmatrix} 2 & 1 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 2v_1 + v_2 \\ 3v_1 + 4v_2 \end{pmatrix}$$

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We want the matrix that makes this true.

$$\operatorname{Rep}_{B,D}(6f) \cdot \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 6(2v_1 + v_2) \\ 6(3v_1 + 4v_2) \end{pmatrix} = \begin{pmatrix} 12v_1 + 6v_2 \\ 18v_1 + 24v_2 \end{pmatrix}$$

Therefore, at least in this example

$$Rep_{B,D}(4f) = \begin{pmatrix} 12 & 6 \\ 18 & 24 \end{pmatrix}$$

so going from the function f to the function 6f has the effect on the representation F of multiplying all the entries by 6.

*Example* Suppose that  $f, g: V \to W$  are linear maps represented with respect to some bases by these matrices.

$$F = \operatorname{Rep}_{B,D}(f) = \begin{pmatrix} 2 & 1 \\ 3 & 4 \end{pmatrix}$$
  $G = \operatorname{Rep}_{B,D}(g) = \begin{pmatrix} 5 & 8 \\ 7 & 6 \end{pmatrix}$ 

We want the matrix representing the function f + g.

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If  $f(\vec{v}) = \vec{u}$  and  $g(\vec{v}) = \vec{w}$  then f + g does this.

$$\vec{v} \stackrel{f+g}{\longmapsto} \vec{u} + \vec{v}$$

$$= u_1 \vec{\delta}_1 + u_2 \vec{\delta}_2 + w_1 \vec{\delta}_1 + w_2 \vec{\delta}_2$$

$$= (u_1 + w_1) \vec{\delta}_1 + (u_2 + w_2) \vec{\delta}_2$$

Thus for any  $\vec{v}$ , the effect on the representatives

$$\operatorname{Rep}_{D}(f(\vec{v})) = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \quad \operatorname{Rep}_{D}(g(\vec{v})) = \begin{pmatrix} w_1 \\ w_2 \end{pmatrix}$$

of adding the functions is to add the column vectors.

$$\operatorname{Rep}_{D}((f+g)(\vec{v})) = \begin{pmatrix} u_1 + w_1 \\ u_2 + w_2 \end{pmatrix}$$

The particular functions in this example have this action on the representations.

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$$\operatorname{Rep}_{D}(g(\vec{v})) = \begin{pmatrix} 5 & 8 \\ 7 & 6 \end{pmatrix} \begin{pmatrix} v_{1} \\ v_{2} \end{pmatrix} = \begin{pmatrix} 5v_{1} + 8v_{2} \\ 7v_{1} + 6v_{2} \end{pmatrix}$$

So we want the matrix with this effect.

$$\operatorname{Rep}_{D}((f+g)(\vec{v}))\begin{pmatrix} v_{1} \\ v_{2} \end{pmatrix} = \begin{pmatrix} 7v_{1} + 9v_{2} \\ 10v_{1} + 10v_{2} \end{pmatrix}$$

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Thus

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and at least in this example the representation of the sum function is the entry-by-entry sum of the representations of the two functions.

#### Definition of matrix sum and scalar multiple

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Example Where

$$A = \begin{pmatrix} 1 & -1 \\ 2 & 3 \end{pmatrix} \quad B = \begin{pmatrix} 0 & 0 & 2 \\ 9 & -1/2 & 5 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0 \\ 8 & -1 \end{pmatrix}$$

Then

$$A + C = \begin{pmatrix} 2 & -1 \\ 10 & 2 \end{pmatrix}$$
  $5B = \begin{pmatrix} 0 & 0 & 10 \\ 45 & -5/2 & 25 \end{pmatrix}$ 

Because the sizes don't match, none of these is defined: A + B, B + A, B + C, C + B.

Proof We can generalize the two examples that started this section. See Exercise 9 . QED

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Example The zero matrix is the identity element for matrix addition.

$$\begin{pmatrix} 3 & 1 & 2 \\ 5 & 0 & 9 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 3 & 1 & 2 \\ 5 & 0 & 9 \end{pmatrix}$$

This reflects the fact that a zero matrix represents a zero function  $Z: V \to W$ , which is the identity elements for function addition.

# Matrix Multiplication

2.1 Lemma The composition of linear maps is linear.

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$$\begin{split} &g \circ h\left(c_1 \cdot \vec{v}_1 + c_2 \cdot \vec{v}_2\right) = g\big(\,h(c_1 \cdot \vec{v}_1 + c_2 \cdot \vec{v}_2)\,\big) = g\big(\,c_1 \cdot h(\vec{v}_1) + c_2 \cdot h(\vec{v}_2)\,\big) \\ &= c_1 \cdot g\big(h(\vec{v}_1)) + c_2 \cdot g(h(\vec{v}_2)\big) = c_1 \cdot (g \circ h)(\vec{v}_1) + c_2 \cdot (g \circ h)(\vec{v}_2) \end{split}$$

shows that  $g \circ h \colon V \to U$  preserves linear combinations. QED

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We will see how the matrix representations of the two functions combine to make the matrix representation of their composition.

*Example* Consider two linear functions h:  $V \to W$  and g:  $W \to X$  represented as here.

$$Rep_{B,C}(h) = \begin{pmatrix} 3 & 1 \\ 2 & 5 \\ 4 & 6 \end{pmatrix}$$
  $Rep_{C,D}(g) = \begin{pmatrix} 8 & 7 & 11 \\ 9 & 10 & 12 \end{pmatrix}$ 

(The sizes of the matrices show that V has dimension 2, W has dimension 3, and X has dimension 2.) We want to see how these two matrices combine to represent the map  $g \circ h: V \to X$ .

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(The sizes of the matrices show that V has dimension 2, W has dimension 3, and X has dimension 2.) We want to see how these two matrices combine to represent the map  $g \circ h: V \to X$ .

We will compute the action of  $g \circ h$  by first computing the action of h on a vector  $\vec{v} \in V$ .

$$\operatorname{Rep}_{C}(h(\vec{v})) = \operatorname{Rep}_{B,C}(h) \cdot \operatorname{Rep}_{B}(\vec{v}) = \begin{pmatrix} 3 & 1 \\ 2 & 5 \\ 4 & 6 \end{pmatrix} \begin{pmatrix} v_{1} \\ v_{2} \end{pmatrix} = \begin{pmatrix} 3v_{1} + v_{2} \\ 2v_{1} + 5v_{2} \\ 4v_{1} + 6v_{2} \end{pmatrix}$$

Apply g to that.

$$Rep_{C,D}(g) \cdot Rep_{C}(h(\vec{v})) = \begin{pmatrix} 8 & 7 & 11 \\ 9 & 10 & 12 \end{pmatrix} \begin{pmatrix} 3\nu_{1} + \nu_{2} \\ 2\nu_{1} + 5\nu_{2} \\ 4\nu_{1} + 6\nu_{2} \end{pmatrix}$$
$$= \begin{pmatrix} 8(3\nu_{1} + \nu_{2}) + 7(2\nu_{1} + 5\nu_{2}) + 11(4\nu_{1} + 6\nu_{2}) \\ 9(3\nu_{1} + \nu_{2}) + 10(2\nu_{1} + 5\nu_{2}) + 12(4\nu_{1} + 6\nu_{2}) \end{pmatrix}$$

Gather terms.

$$= \begin{pmatrix} (8 \cdot 3 + 7 \cdot 2 + 11 \cdot 4)\nu_1 + (8 \cdot 1 + 7 \cdot 5 + 11 \cdot 6)\nu_2 \\ (9 \cdot 3 + 10 \cdot 2 + 12 \cdot 4)\nu_1 + (9 \cdot 1 + 10 \cdot 5 + 12 \cdot 6)\nu_2 \end{pmatrix}$$

Rewrite that as a matrix-vector multiplication.

$$\begin{pmatrix} 8 \cdot 3 + 7 \cdot 2 + 11 \cdot 4 & 8 \cdot 1 + 7 \cdot 5 + 11 \cdot 6 \\ 9 \cdot 3 + 10 \cdot 2 + 12 \cdot 4 & 9 \cdot 1 + 10 \cdot 5 + 12 \cdot 6 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

Here is the combination.

$$\begin{pmatrix} 8 & 7 & 11 \\ 9 & 10 & 12 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 2 & 5 \\ 4 & 6 \end{pmatrix} = \begin{pmatrix} 8 \cdot 3 + 7 \cdot 2 + 11 \cdot 4 & 8 \cdot 1 + 7 \cdot 5 + 11 \cdot 6 \\ 9 \cdot 3 + 10 \cdot 2 + 12 \cdot 4 & 9 \cdot 1 + 10 \cdot 5 + 12 \cdot 6 \end{pmatrix}$$

#### Definition of matrix multiplication

2.3 Definition The matrix-multiplicative product of the  $m \times r$  matrix G and the  $r \times n$  matrix H is the  $m \times n$  matrix P, where

$$p_{\mathfrak{i},\mathfrak{j}} = g_{\mathfrak{i},1}h_{1,\mathfrak{j}} + g_{\mathfrak{i},2}h_{2,\mathfrak{j}} + \cdots + g_{\mathfrak{i},r}h_{r,\mathfrak{j}}$$

that is, the i, j-th entry of the product is the dot product of the i-th row of the first matrix with the j-th column of the second.

$$\mathsf{GH} = \begin{pmatrix} & \vdots & & \\ g_{\mathfrak{i},1} & g_{\mathfrak{i},2} & \dots & g_{\mathfrak{i},r} \\ & \vdots & & & \end{pmatrix} \begin{pmatrix} & h_{1,\mathfrak{j}} \\ \dots & h_{2,\mathfrak{j}} & \dots \\ & \vdots \\ & h_{r,\mathfrak{j}} & \end{pmatrix} = \begin{pmatrix} & \vdots \\ \dots & p_{\mathfrak{i},\mathfrak{j}} & \dots \\ & \vdots & & \\ & \vdots & & \end{pmatrix}$$

Example

$$\begin{pmatrix} 3 & 1 & 6 \\ 2 & 5 & 9 \end{pmatrix} \begin{pmatrix} 2 & 0 & 4 \\ 1 & -3 & 5 \\ 4 & 2 & 7 \end{pmatrix} = \begin{pmatrix} 31 & 9 & 59 \\ 45 & 3 & 96 \end{pmatrix}$$

#### Example This product is not defined.

$$\begin{pmatrix} 1 & 3 & -1 \\ 0 & 0 & 0 \\ 2 & 0 & 0 \end{pmatrix} \begin{pmatrix} 5 & 7 & 1 \\ 2 & 2 & 0 \end{pmatrix}$$

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Example Square matrices of the same size have a defined product.

$$\begin{pmatrix} 1 & 3 & -1 \\ 0 & 0 & 0 \\ 2 & 0 & 0 \end{pmatrix} \begin{pmatrix} 5 & 7 & 1 \\ 2 & 2 & 0 \\ 1 & -1 & 2 \end{pmatrix} = \begin{pmatrix} 10 & 14 & -1 \\ 0 & 0 & 0 \\ 10 & 14 & 2 \end{pmatrix}$$

This reflects the fact that we can compose two functions from a space to itself  $f, g: V \to V$ .

#### Matrix multiplication represents composition

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**Proof** Let h:  $V \to W$  and g:  $W \to X$  be represented by H and G with respect to bases  $B \subset V$ ,  $C \subset W$ , and  $D \subset X$ , of sizes n, r, and m. For any  $\vec{v} \in V$ , the k-th component of  $\operatorname{Rep}_C(h(\vec{v}))$  is

$$h_{k,1}v_1 + \cdots + h_{k,n}v_n$$

and so the i-th component of  $Rep_{D}(g \circ h(\vec{v}))$  is this.

$$\begin{split} g_{i,1} \cdot (h_{1,1}\nu_1 + \dots + h_{1,n}\nu_n) + g_{i,2} \cdot (h_{2,1}\nu_1 + \dots + h_{2,n}\nu_n) \\ + \dots + g_{i,r} \cdot (h_{r,1}\nu_1 + \dots + h_{r,n}\nu_n) \end{split}$$

Distribute and regroup on the v's.

$$= (g_{i,1}h_{1,1} + g_{i,2}h_{2,1} + \dots + g_{i,r}h_{r,1}) \cdot \nu_1$$
  
 
$$+ \dots + (g_{i,1}h_{1,n} + g_{i,2}h_{2,n} + \dots + g_{i,r}h_{r,n}) \cdot \nu_n$$

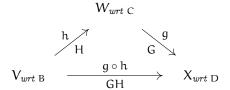
Finish by recognizing that the coefficient of each  $v_i$ 

$$g_{i,1}h_{1,j} + g_{i,2}h_{2,j} + \cdots + g_{i,r}h_{r,j}$$

matches the definition of the i,j entry of the product GH. QED

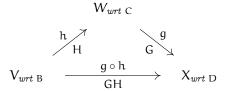
#### Arrow diagrams

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Above the arrows, the maps show that the two ways of going from V to X, straight over via the composition or else in two steps by way of W, have the same effect

$$\vec{v} \overset{g \circ h}{\longmapsto} g(h(\vec{v})) \qquad \vec{v} \overset{h}{\longmapsto} h(\vec{v}) \overset{g}{\longmapsto} g(h(\vec{v}))$$

(this is just the definition of composition). Below the arrows, the matrices indicate that the product does the same thing—multiplying GH into the column vector  $\operatorname{Rep}_B(\vec{\nu})$  has the same effect as multiplying the column vector first by H and then multiplying the result by G.

$$Rep_{B,D}(g \circ h) = GH$$
  $Rep_{C,D}(g) Rep_{B,C}(h) = GH$ 

#### Order, dimensions, and sizes

Consider the composition of two function.

First consider the order in which we write the composition. In  $g \circ h$  the function written first, g, is the function applied second.

$$\vec{v} \stackrel{h}{\longmapsto} h(\vec{v}) \stackrel{g}{\longmapsto} g(h(\vec{v}))$$

We write g first to match the definition  $g \circ h(\vec{v}) = g(h(\vec{v}))$ . That order carries over to matrices:  $g \circ h$  is represented by GH.

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Now consider the dimensions of the domain and codomain spaces.

 $\text{dimension } n \text{ space } \stackrel{h}{\longrightarrow} \text{ dimension } r \text{ space } \stackrel{g}{\longrightarrow} \text{ dimension } \mathfrak{m} \text{ space }$ 

The representing  $m \times n$  matrix GH is the product of an  $m \times r$  matrix G and a  $r \times n$  matrix H. Briefly, ' $m \times r$  times  $r \times n$  equals  $m \times n$ '.

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Function composition is not a commutative operation —  $\cos(x^2)$  is different than  $\cos^2(x)$ . This also holds in the special case of composition of linear functions.

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Example Changing the order in which we multiply these matrices

$$\begin{pmatrix} 3 & 3 \\ 0 & 4 \end{pmatrix} \begin{pmatrix} -2 & 6 \\ 6 & 5 \end{pmatrix} = \begin{pmatrix} 12 & 33 \\ 24 & 20 \end{pmatrix}$$

changes the result.

$$\begin{pmatrix} -2 & 6 \\ 6 & 5 \end{pmatrix} \begin{pmatrix} 3 & 3 \\ 0 & 4 \end{pmatrix} = \begin{pmatrix} -6 & 18 \\ 18 & 38 \end{pmatrix}$$

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Example The product of these matrices

$$\begin{pmatrix} 3 & 4 \\ 0 & 2 \end{pmatrix} \qquad \begin{pmatrix} 8 & 12 & 0 \\ -4 & 0 & 1/2 \end{pmatrix}$$

is defined in one order and not defined in the other.

2.12 Theorem If F, G, and H are matrices, and the matrix products are defined, then the product is associative (FG)H = F(GH) and distributes over matrix addition F(G+H) = FG + FH and (G+H)F = GF + HF.

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**Proof** Associativity holds because matrix multiplication represents function composition, which is associative: the maps  $(f \circ g) \circ h$  and  $f \circ (g \circ h)$  are equal as both send  $\vec{v}$  to  $f(g(h(\vec{v})))$ .

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Distributivity is similar. For instance, the first one goes  $f\circ (g+h)\,(\vec{v})=f\big(\,(g+h)(\vec{v})\,\big)=f\big(\,g(\vec{v})+h(\vec{v})\,\big)=f(g(\vec{v}))+f(h(\vec{v}))=f\circ g(\vec{v})+f\circ h(\vec{v})\,\big)$  (the third equality uses the linearity of f). QED

Mechanics of Matrix Multiplication

#### Combinatorics of multiplication

The striking thing about matrix multiplication is the way rows and columns combine. Here a second row and a third column combine to make a 2,3 entry.

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 4 & 6 & 8 \\ 5 & 7 & 9 \end{pmatrix} = \begin{pmatrix} 9 & 13 & 17 \\ 5 & 7 & 9 \\ 4 & 6 & 8 \end{pmatrix}$$

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The i, j entry of the matrix product GH is the dot product of row i of the left matrix G with column j of the right one H.

$$p_{i,j} = g_{i, ||} h_{|||,j} + g_{i, ||} h_{|||,j} + \cdots + g_{i, ||} h_{|||,j} + \cdots + g_{i, ||} h_{|||,j}$$

## Combinatorics of multiplication

The striking thing about matrix multiplication is the way rows and columns combine. Here a second row and a third column combine to make a 2,3 entry.

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 4 & 6 & 8 \\ 5 & 7 & 9 \end{pmatrix} = \begin{pmatrix} 9 & 13 & 17 \\ 5 & 7 & 9 \\ 4 & 6 & 8 \end{pmatrix}$$

The i, j entry of the matrix product GH is the dot product of row i of the left matrix G with column j of the right one H.

$$p_{i,j} = g_{i, |||} h_{|||,j} + g_{i, |||} h_{|||,j} + \cdots + g_{i, |||} h_{|||,j} + \cdots + g_{i, |||} h_{||||,j}$$

We can view this as the left matrix acting by multiplying its rows into the columns of the right matrix. Or we could see it as the right matrix using its columns to act on the left matrix's rows.

3.2 Definition A matrix with all 0's except for a 1 in the i, j entry is an i, j unit matrix.

*Example* The 2,1 unit  $2\times3$  matrix multiplies from the left

$$\begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 2 & 1 & 3 \\ 5 & 6 & 4 \\ 7 & 8 & 9 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 2 & 1 & 3 \\ 0 & 0 & 0 \end{pmatrix}$$

to copy row 1 of the multiplicand into row 2 of the result.

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*Example* From the right the 2, 1 unit  $2\times3$  matrix

$$\begin{pmatrix} 3 & 4 \\ 6 & 5 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 0 & 0 \\ 5 & 0 & 0 \end{pmatrix}$$

copies column 2 of the first matrix into column 1 of the result.

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copies column 2 of the first matrix into column 1 of the result. Example Rescaling the unit matrix rescales the result.

$$\begin{pmatrix} 3 & 4 \\ 6 & 5 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 \\ 3 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 12 & 0 & 0 \\ 15 & 0 & 0 \end{pmatrix}$$

3.8 Lemma In a product of two matrices G and H, the columns of GH are formed by taking G times the columns of H

$$G \cdot \begin{pmatrix} \vdots \\ \vec{h}_1 \\ \vdots \end{pmatrix} \cdots \begin{pmatrix} \vdots \\ \vec{h}_n \\ \vdots \end{pmatrix} = \begin{pmatrix} \vdots \\ G \cdot \vec{h}_1 \\ \vdots \end{pmatrix} \cdots \begin{pmatrix} \vdots \\ G \cdot \vec{h}_n \\ \vdots \end{pmatrix}$$

and the rows of GH are formed by taking the rows of G times H

$$\left( \frac{\cdots \vec{g}_1 \cdots}{\vdots} \right) \cdot H = \left( \frac{\cdots \vec{g}_1 \cdot H \cdots}{\vdots} \right)$$

$$\cdots \vec{g}_r \cdots \right)$$

(ignoring the extra parentheses).

**Proof** We will check that in a product of  $2\times 2$  matrices, the rows of the product equal the product of the rows of G with the entire matrix H.

$$\begin{pmatrix} g_{1,1} & g_{1,2} \\ g_{2,1} & g_{2,2} \end{pmatrix} \begin{pmatrix} h_{1,1} & h_{1,2} \\ h_{2,1} & h_{2,2} \end{pmatrix} = \begin{pmatrix} (g_{1,1} & g_{1,2})H \\ (g_{2,1} & g_{2,2})H \end{pmatrix}$$
 
$$= \begin{pmatrix} (g_{1,1}h_{1,1} + g_{1,2}h_{2,1} & g_{1,1}h_{1,2} + g_{1,2}h_{2,2}) \\ (g_{2,1}h_{1,1} + g_{2,2}h_{2,1} & g_{2,1}h_{1,2} + g_{2,2}h_{2,2}) \end{pmatrix}$$

We will leave the more general check as an exercise. QED

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- 3.10 Definition An identity matrix is square and every entry is 0 except for 1's in the main diagonal.

$$I_{n\times n} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ & \vdots & & \\ 0 & 0 & \dots & 1 \end{pmatrix}$$

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Taking the product with an identity matrix returns the multiplicand. Example Multiplication by an identity from the left

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix} = \begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix}$$

or from the right leaves the matrix unchanged.

$$\begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix}$$

3.13 *Definition* diagonal.

A diagonal matrix is square and has 0's off the main

$$\begin{pmatrix} a_{1,1} & 0 & \dots & 0 \\ 0 & a_{2,2} & \dots & 0 \\ & \vdots & & & \\ 0 & 0 & \dots & a_{n,n} \end{pmatrix}$$

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*Example* Multiplication from the left by a diagonal matrix rescales the rows.

$$\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix} = \begin{pmatrix} 6 & 4 \\ -3 & 15 \end{pmatrix}$$

From the right it rescales the columns.

$$\begin{pmatrix} 3 & 2 \\ -1 & 5 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} = \begin{pmatrix} 6 & 6 \\ -2 & 15 \end{pmatrix}$$

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3.15 *Definition* A *permutation matrix* is square and is all 0's except for a single 1 in each row and column.

*Example* Multiplication by a permutation matrix from the left will swap rows.

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} = \begin{pmatrix} 4 & 5 & 6 \\ 1 & 2 & 3 \\ 7 & 8 & 9 \end{pmatrix}$$

From the right it swaps columns.

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \quad \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1 & 3 \\ 5 & 4 & 6 \\ 8 & 7 & 9 \end{pmatrix}$$

3.19 *Definition* The *elementary reduction matrices* result from applying a one Gaussian operation to an identity matrix.

- 1.  $I \xrightarrow{k\rho_i} M_i(k)$  for  $k \neq 0$
- 2. I  $\stackrel{\rho_i \leftrightarrow \rho_j}{\longrightarrow} P_{i,j}$  for  $i \neq j$
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*Example* Here are some  $2 \times 2$  examples.

$$M_2(3) = \begin{pmatrix} 1 & 0 \\ 0 & 3 \end{pmatrix} \quad P_{1,2} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad C_{1,2}(-3) = \begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix}$$

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*Example* Some  $3 \times 3$  examples.

$$P_{2,3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \quad C_{2,3}(-4) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -4 & 0 \end{pmatrix}$$

*Example* Multiplying on the left by the  $3\times3$  matrix  $M_2(1/2)$  has the effect of the row operation  $(1/2)\rho_2$ .

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 3 & 0 \\ 0 & 2 & 5 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 3 & 0 \\ 0 & 1 & 5/2 \\ 0 & 0 & 0 \end{pmatrix}$$

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Example Left multiplication by  $C_{1,3}(-2)$  performs the row operation  $-2\rho_1 + \rho_3$ .

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 3 & 0 \\ 0 & 2 & 5 \\ 2 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 3 & 0 \\ 0 & 2 & 5 \\ 0 & -5 & 1 \end{pmatrix}$$

- 3.20 *Lemma* Gaussian reduction can be done through matrix multiplication.
  - 1. If  $H \xrightarrow{k\rho_i} G$  then  $M_i(k)H = G$ .
  - 2. If  $H \stackrel{\rho_i \leftrightarrow \rho_j}{\longrightarrow} G$  then  $P_{i,j}H = G$ .
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Proof Clear.

QED

3.23 Corollary For any matrix H there are elementary reduction matrices  $R_1, \ldots, R_r$  such that  $R_r \cdot R_{r-1} \cdots R_1 \cdot H$  is in reduced echelon form.

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*Example* We can bring this augmented matrix to echelon form with matrix multiplication.

$$\begin{pmatrix} 1 & -1 & 2 & | & 4 \\ 2 & -2 & -1 & | & 6 \\ 0 & 3 & 1 & | & 5 \end{pmatrix}$$

We first perform  $-2\rho_1+\rho_2$  via left multiplication by  $C_{1,2}(-2).$ 

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Now we swap rows 2 and 3 with  $P_{2,3}$ 

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When writing them out in full, remember that the matrix used first appears to the right of the matrix used second.

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 2 & | & 4 \\ 2 & -2 & -1 & | & 6 \\ 0 & 3 & 1 & | & 5 \end{pmatrix} = \begin{pmatrix} 1 & -1 & 2 & | & 4 \\ 0 & 3 & 1 & | & 5 \\ 0 & 0 & -5 & | & -2 \end{pmatrix}$$

# Inverses

#### Function inverses

We finish this section by considering how to represent the inverse of a linear map.

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We first recall some things about inverses. Where  $\pi\colon\mathbb{R}^3\to\mathbb{R}^2$  is the projection map and  $\iota\colon\mathbb{R}^2\to\mathbb{R}^3$  is the embedding

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \stackrel{\pi}{\longmapsto} \begin{pmatrix} x \\ y \end{pmatrix} \qquad \begin{pmatrix} x \\ y \end{pmatrix} \stackrel{\iota}{\longmapsto} \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$$

then the composition  $\pi \circ \iota$  is the identity map  $\pi \circ \iota = id$  on  $\mathbb{R}^2$ .

$$\begin{pmatrix} x \\ y \end{pmatrix} \stackrel{\iota}{\longmapsto} \begin{pmatrix} x \\ y \\ 0 \end{pmatrix} \stackrel{\pi}{\longmapsto} \begin{pmatrix} x \\ y \end{pmatrix}$$

We say that  $\iota$  is a *right inverse* of  $\pi$  or, what is the same thing, that  $\pi$  is a *left inverse* of  $\iota$ .

However, composition in the other order  $\iota \circ \pi$  doesn't give the identity map—here is a vector that is not sent to itself under  $\iota \circ \pi$ .

$$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \stackrel{\pi}{\longmapsto} \begin{pmatrix} 0 \\ 0 \end{pmatrix} \stackrel{\iota}{\longmapsto} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

However, composition in the other order  $\iota \circ \pi$  doesn't give the identity map—here is a vector that is not sent to itself under  $\iota \circ \pi$ .

$$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \stackrel{\pi}{\longmapsto} \begin{pmatrix} 0 \\ 0 \end{pmatrix} \stackrel{\iota}{\longmapsto} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

In fact,  $\pi$  has no left inverse at all. For, if f were to be a left inverse of  $\pi$  then we would have

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \stackrel{\pi}{\longmapsto} \begin{pmatrix} x \\ y \end{pmatrix} \stackrel{f}{\longmapsto} \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

for all of the infinitely many z's. But no function can send a single argument to more than one value. (An example of a function with no inverse on either side is the zero transformation on  $\mathbb{R}^2$ .)

Some functions have a *two-sided inverse*, another function that is the inverse of the first both from the left and from the right. For instance, the map given by  $\vec{v}\mapsto 2\cdot\vec{v}$  has the two-sided inverse  $\vec{v}\mapsto (1/2)\cdot\vec{v}$ . The appendix shows that a function has a two-sided inverse if and only if it is both one-to-one and onto. The appendix also shows that if a function f has a two-sided inverse then it is unique, and so we call it 'the' inverse and denote it  $f^{-1}$ .

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In addition, recall that we have shown in Theorem II.2.21 that if a linear map has a two-sided inverse then that inverse is also linear.

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In addition, recall that we have shown in Theorem II.2.21 that if a linear map has a two-sided inverse then that inverse is also linear.

Thus, our goal in this subsection is, where a linear h has an inverse, to find the relationship between  $Rep_{B,D}(h)$  and  $Rep_{D,B}(h^{-1})$ .

## Definition of matrix inverse

4.1 Definition A matrix G is a left inverse matrix of the matrix H if GH is the identity matrix. It is a right inverse matrix if HG is the identity. A matrix H with a two-sided inverse is an invertible matrix. That two-sided inverse is the inverse matrix and is denoted H<sup>-1</sup>.

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Example This matrix

$$H = \begin{pmatrix} 2 & 5 \\ 1 & 3 \end{pmatrix}$$

has a two-sided inverse.

$$H^{-1} = \begin{pmatrix} 3 & -5 \\ -1 & 2 \end{pmatrix}$$

To check that we can multiply in both orders. Here is one order; the other is just as easy to check.

$$\begin{pmatrix} 2 & 5 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} 3 & -5 \\ -1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

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Proof (For both results.) Given a matrix H, fix spaces of appropriate dimension for the domain and codomain. Fix bases for these spaces. With respect to these bases, H represents a map h. The statements are true about the map and therefore they are true about the matrix.

*Proof* Because the two matrices are invertible they are square. Because their product is defined they must be square of the same dimension,  $n \times n$ . So by fixing a basis for  $\mathbb{R}^n$ —we can use the standard basis—we get maps  $g,h\colon \mathbb{R}^n\to \mathbb{R}^n$  that are associated with the matrices,  $G=\operatorname{Rep}_{\mathcal{E}_n,\mathcal{E}_n}(g)$  and  $H=\operatorname{Rep}_{\mathcal{E}_n,\mathcal{E}_n}(h)$ .

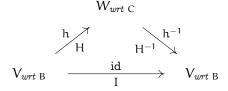
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Consider  $h^{-1}g^{-1}$ . By the prior paragraph this composition is defined. This map is a two-sided inverse of gh since  $(h^{-1}g^{-1})(gh) = h^{-1}(id)h = h^{-1}h = id$  and  $(gh)(h^{-1}g^{-1}) = g(id)g^{-1} = gg^{-1} = id$ . the matrices representing the maps reflect this equality. QED

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Here is the arrow diagram for matrix inverses.



**Proof** The matrix H is invertible if and only if it is nonsingular and thus Gauss-Jordan reduces to the identity. By Corollary 3.23 we can do this reduction with elementary matrices.

$$R_r \cdot R_{r-1} \dots R_1 \cdot H = I \tag{*}$$

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For the first sentence of the result, note that elementary matrices are invertible (because elementary row operations are reversible) and that their inverses are also elementary. Apply  $R_r^{-1}$  from the left to both sides of (\*). Then apply  $R_{r-1}^{-1}$ , etc. The result gives H as the product of elementary matrices  $H = R_1^{-1} \cdots R_r^{-1} \cdot I$  (the I here covers the trivial r = 0 case).

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For the second sentence, rewrite (\*) as  $(R_r \cdot R_{r-1} \dots R_1) \cdot H = I$  to recognize that  $H^{-1} = R_r \cdot R_{r-1} \dots R_1 \cdot I$ . Restated, applying  $R_1$  to the identity, followed by  $R_2$ , etc., yields the inverse of H. QED

Example This matrix is nonsingular and so invertible.

$$A = \begin{pmatrix} 1 & 3 & 1 \\ 2 & 0 & -1 \\ 1 & 2 & 0 \end{pmatrix}$$

To ease the calculation called for in the Lemma, we write the matrix A next to the  $3\times3$  identity and as we Gauss-Jordan reduce A we apply those operations to I.

$$\begin{pmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
2 & 0 & -1 & 0 & 1 & 0 \\
1 & 2 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{-2\rho_1 + \rho_2}
\begin{pmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & -6 & -3 & -2 & 1 & 0 \\
0 & -1 & -1 & -1 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{-1/6\rho_2 + \rho_3}
\begin{pmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & -6 & -3 & -2 & 1 & 0 \\
0 & 0 & -1/2 & -2/3 & -1/6 & 1
\end{pmatrix}$$

$$\xrightarrow{-1/6\rho_2}
\begin{pmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & 0 & -1/2 & 1/3 & -1/6 & 0 \\
0 & 0 & 1 & 1/2 & 1/3 & -1/6 & 0 \\
0 & 0 & 1 & 4/3 & 1/3 & -2
\end{pmatrix}$$

$$\begin{pmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & 1 & 1/2 & 1/3 & -1/6 & 0 \\
0 & 0 & 1 & 4/3 & 1/3 & -2
\end{pmatrix}
\xrightarrow{\begin{array}{c}
-\rho_3 + \rho_1 \\
-(1/2)\rho_3 + \rho_2
\end{array}}
\begin{pmatrix}
1 & 3 & 0 & -1/3 & -1/3 & 2 \\
0 & 1 & 0 & -1/3 & -1/3 & 1 \\
0 & 0 & 1 & 4/3 & 1/3 & -2
\end{pmatrix}$$

$$\xrightarrow{\begin{array}{c}
-3\rho_2 + \rho_1 \\
\rightarrow
\end{array}}
\begin{pmatrix}
1 & 0 & 0 & 2/3 & 2/3 & -1 \\
0 & 1 & 0 & -1/3 & -1/3 & 1 \\
0 & 0 & 1 & 4/3 & 1/3 & -2
\end{pmatrix}$$

We have calculated that this is the inverse.

$$A^{-1} = \begin{pmatrix} 2/3 & 2/3 & -1 \\ -1/3 & -1/3 & 1 \\ 4/2 & 1/2 & 2 \end{pmatrix}$$

Finding the inverse of a matrix A is a lot of work but once we have it then solving linear systems  $A\vec{x} = \vec{b}$  is easy.

*Example* The linear system

$$x + 3y + z = 2$$

$$2x - z = 12$$

$$x + 2y = 4$$

is this matrix equation.

$$\begin{pmatrix} 1 & 3 & 1 \\ 2 & 0 & -1 \\ 1 & 2 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2 \\ 12 \\ 4 \end{pmatrix}$$

Solve it by multiplying both sides from the left by the inverse that we found earlier.

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2/3 & 2/3 & -1 \\ -1/3 & -1/3 & 1 \\ 4/3 & 1/3 & -2 \end{pmatrix} \begin{pmatrix} 2 \\ 12 \\ 4 \end{pmatrix} = \begin{pmatrix} 16/3 \\ -2/3 \\ -4/3 \end{pmatrix}$$

4.11 Corollary The inverse for a  $2\times 2$  matrix exists and equals

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

if and only if  $ad - bc \neq 0$ .

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The  $3\times3$  formula is much more complicated. We will cover it in the next chapter.